

Tackling the problem of antimicrobial resistance



Collaboration between Singapore
General Hospital and DXC Technology

Overuse or misuse of antibiotics in hospital settings is a global problem, fueling one of the most serious public health threats: antimicrobial resistance. An estimated 20 per cent to 50 per cent of acute care hospitals worldwide aren't prescribing antibiotics appropriately. As a result, micro-organisms become resistant to antimicrobials, including antibiotics, and can become superbugs that don't respond to any drugs.

It is believed that up to 30 per cent of infections in hospitals in Singapore are resistant to third-generation cephalosporins, which are widely used broad-spectrum antibiotics. Reviews of Singapore hospitals have found that heavy use of broad-spectrum antibiotics has resulted in more resistance of Gram-negative bacteria. Further compounding the problem is a dwindling antibiotic pipeline, with limited therapeutics for the treatment of multi-drug-resistant organisms (MDROs).

One key strategy to tackle the problem of antimicrobial resistance (AMR) is the practice of antimicrobial stewardship programs (ASP). These initiatives seek to prevent the overuse of antimicrobial agents and ensure optimal selection of antibiotics, for example, where appropriate, making recommendations for the use of narrower-spectrum antibiotics and shorter durations of antibiotic use.

ASP teams have become well established in Singapore's public hospitals, and studies have shown they have been instrumental in improving antibiotic abuse. Crucially, their recommendations have led to shorter hospital stays, a reduction in mortality and readmission rates, and large cost savings for patients. High prescription rates of broad-spectrum antibiotics are understandable since these are often seen as the easiest way to address infection. However, it's important that good strategies are established to guide the use of antibiotics in a more precise and medically appropriate way.

Prescribing methodology

A key priority at acute care hospitals in Singapore, led by ASP teams and supported by precise, comprehensive data, is to establish a clear prescribing methodology.

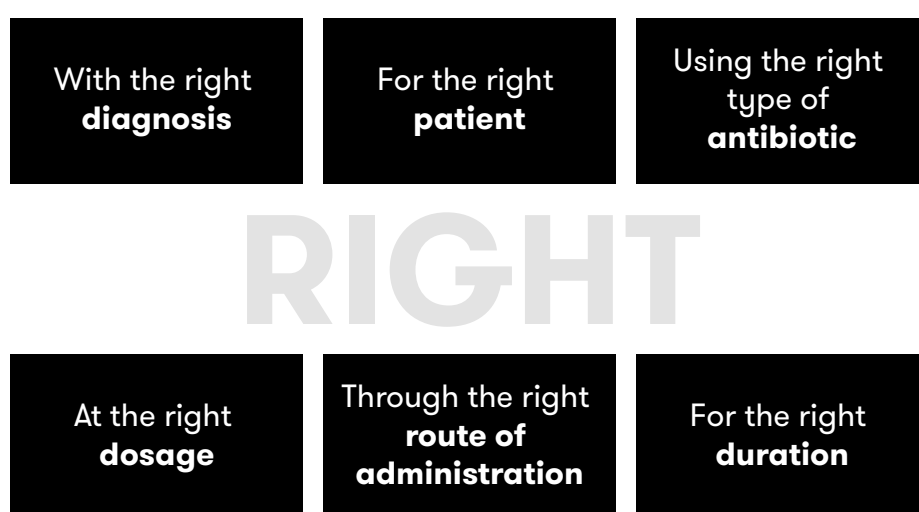


Figure 1. Prescribing methodology: Six key steps

The best approach for ensuring this prescribing methodology will be put into practice is using automation and artificial intelligence (AI) to collate and analyse existing data. These technologies can significantly improve the ability of Singapore hospitals and the broader Singapore healthcare system to alleviate the overuse of antibiotics.

Antimicrobial efforts at Singapore General Hospital

One hospital that has a long-established ASP team is Singapore General Hospital (SGH), and studies have shown its recommended interventions have benefited both patients and the hospital. Nevertheless, problems remain. In 2018 alone, approximately 12,000 audits performed for broad-spectrum antibiotic use at SGH showed that 20 per cent to 30 per cent of these prescriptions were inappropriate. Yet this high volume of audits covered only seven intravenous antibiotics in the 2018 workflow — a small fraction of the range of antibiotics used in the acute hospital setting and in the wider community.

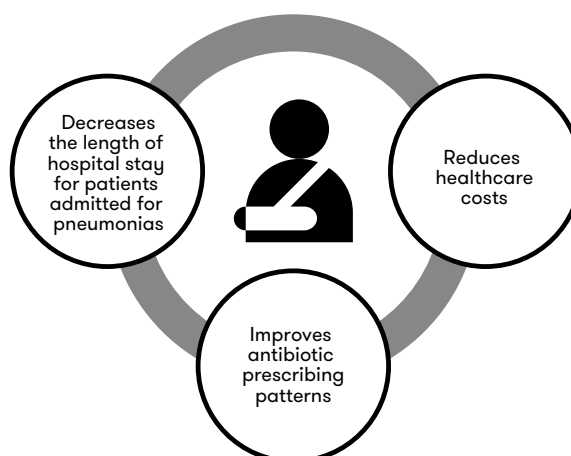
To achieve comprehensive antibiotic stewardship, the ASP team would need to manually audit a wider range of antibiotics — both broad and narrow spectrum delivered intravenously or through tablet form. This manual effort would exceed the manpower capacity of the ASP team.

Beyond simply auditing antibiotic use, the overarching goal is to shift the focus to an assessment of the clinical syndrome or type of infection, thereby adopting a more holistic and pre-emptive strategy to stewardship. This approach would typically include helping the prescribers systemically evaluate their patients' presenting signs and symptoms as well as risk factors that impact antibiotic use. This would be a significant change from current practice (reactive strategy), where typically a patient will have been on an antibiotic for 3 days before the information needed to determine the right antibiotic for that patient has been collated.

A good starting point for a pre-emptive strategy is to focus on the most common type of infection treated at the hospital, pneumonia. According to SGH's own studies, approximately 20 per cent of all infections treated in the hospital setting are due to pneumonia, and more antibiotic prescriptions are written for pneumonia than any other condition. However, not all pneumonias are due to bacterial infections. A proportion of pneumonias are due to viruses, for which antibiotics are not indicated. In addition, not all antibiotics need to be administered intravenously. In some cases, oral antibiotics are appropriate.

This underscores the need for a more patient-centred approach to be considered in stewardship efforts.

Figure 2. Benefits of patient-centred approach



For example, the cost of treating pneumonia in a private hospital is more than S\$8,500 and approximately S\$1,250 to S\$5,000 in public hospitals for government-subsidised patients, with average length of stay of between 2 and 9 days. The potential for improved care and healthcare cost reduction is significant.

The ASP model: Pinpointing the right treatment

To tackle antimicrobial resistance and ensure patients get the appropriate treatment, Singapore General Hospital's ASP is developing a clinical reasoning framework for prescriptions based on five categories of questions, each with markers to assist with the diagnosis and selection of antibiotics:

1. Is there an infection?

This includes whether the patient has a fever, the patient's hemodynamic status and findings from blood tests, such as white blood count, neutrophil count, monocyte count, sputum cultures, throat swab for respiratory viral panel and methicillin-resistant *Staphylococcus aureus* (MRSA) screening. With pneumonia, for example, tests would determine whether the infection was viral versus bacterial.

2. Where is the infection?

What are the presenting complaints? What are the biomarkers from lab tests, x-rays, etc.?

3. When did the infection occur? Has it occurred less than 48 hours ago or greater than 48 hours?

This will help provide insight into whether it is a community-acquired infection or a hospital-acquired infection.

4. Who is the host? What are the medical comorbidities of the patient (using the medical components within the Charlson comorbidity index)? Has the patient had a separate hospitalisation within 90 days? And does the patient have any allergies?

5. What antibiotic should be selected?

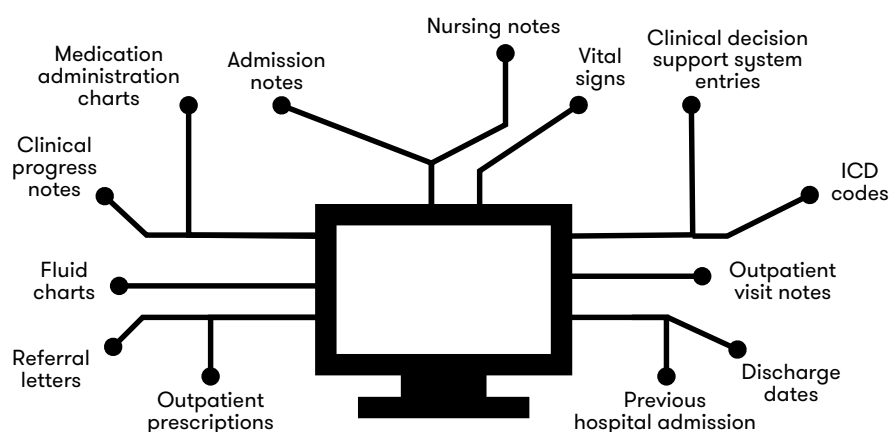
The previous four questions help to narrow down the selection of antibiotics. Further questions on the selection of treatment include: whether the patient has systemic inflammatory response syndrome (SIRS) or septic shock, has been in the ICU, has taken antibiotics in the past 30 days and any culture sensitivity reports concerning the bug and what it might be sensitive to.

This framework can help physicians achieve the objective of arriving at the right diagnosis. If the use of antibiotics is indicated, accurate data allows clinicians to select the narrowest spectrum antibiotic for the treatment of the infection for the right duration. The challenge for the ASP team has been to find a solution that enables them to access, leverage, analyse and act upon this vast amount of data.

To properly determine the type of infection and improve prescribing patterns, the ASP team will need to draw insights from multiple systems of record — first within the institution, then subsequently from national databases if possible.

SGH Institution's proprietary EMR system

Figure 3. An overview of SGH Institution's proprietary EMR system



High cost of antimicrobial review

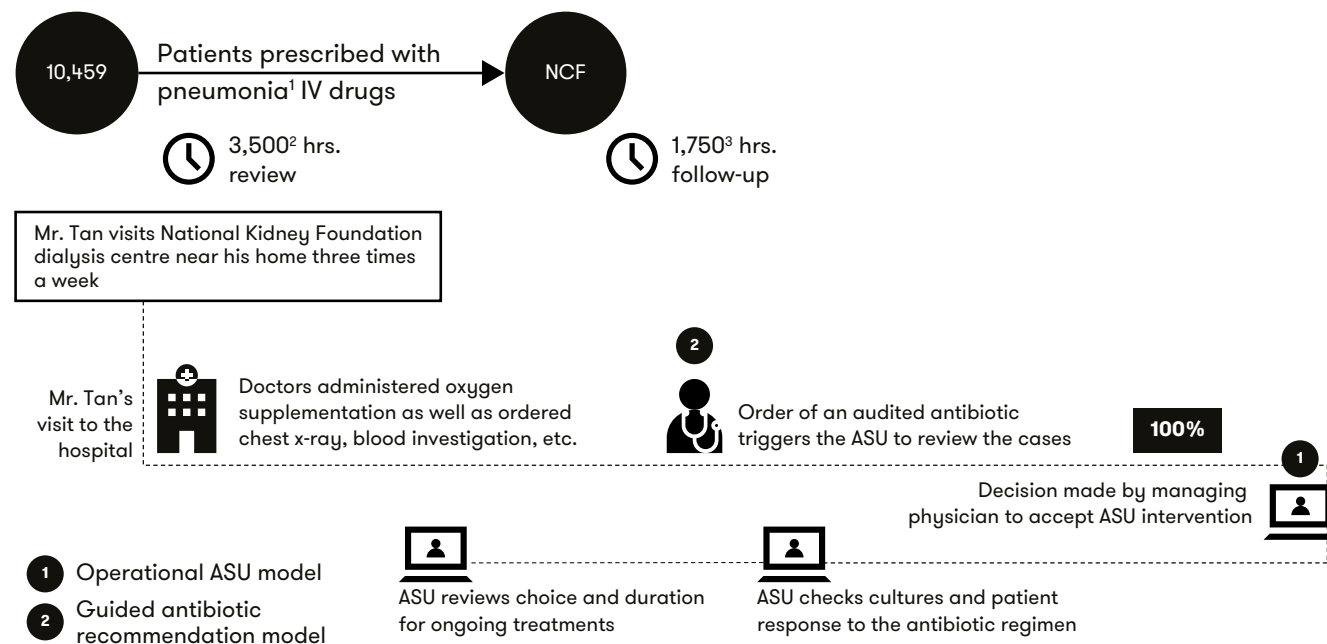


Figure 4. Background: High cost of antimicrobial review

¹ Audited antibiotics covered in the scope of the POC Piperacillin/tazobactam, Ciprofloxacin Injections, Doripenem Injections, Ertapenem Injections, Tienam Injections, Levofloxacin injection and Meropenem Injections

² 20 minutes for each prescription review and 10 minutes for follow-up

³ Based on the number of follow-up patients with pneumonia-prescribed antibiotics within the analyzed period between January 1, 2016 and December 31, 2018

At SGH, data will need to be drawn from the institution's proprietary electronic medical record system. Information from linked data repositories, such as radiology images, all laboratory results, histology reports and anaesthetic charts, is also required to support a syndromic approach for ASP.

This is a vast ecosystem to navigate. To prevent the inappropriate use of antibiotics, the ASP team as well as clinicians would need to have these insights in real time at the point of prescription.

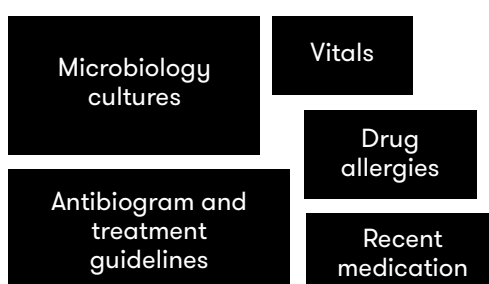
By using AI to identify patient-cum-treatment profiles that require further review, the ASP team would be able to better prioritise complex cases, perform a more thorough analysis of each case prior to making recommendations on antibiotic choice and duration. Based on these recommendations, clinicians could then treat patients appropriately from the outset, and reduce the over-prescribing of broad-spectrum antibiotics.

Automation would also enable the hospital to expand its audit coverage to a larger number of antibiotics prescribed in the hospital, thus increasing the service's productivity, and provide more thorough insight into the institution's antibiotic consumption.

Managing a complex data model

Over the past 4 months, DXC and SGH have been co-creating an artificial intelligence model that will enable faster, more accurate insights into antibiotic use and misuse once certain data challenges are overcome.

Figure 5. Challenges with unstandardized data



Vital information needed for ASPs to review antibiotic use

The information stored in the hospital's complex ecosystem comprises a mix of structured, semi-structured and unstructured data.

The data needed are not standardised or in a clear format.

The teams developing the AI model must address problems of data quality, data engineering, data freshness and data completeness. The table below shows the disparate information within various systems.

Data	Source
Vitals (blood pressure, heart rate, oxygen saturation, temperature)	Sunrise clinical manager (SCM)
Organ function (biochemistry results, fluid charts)	SCM
Drug allergies, pregnancy status	SCM/National Electronic Health Records (NEHR)
Post positive microbiology cultures (recent 3 months)	SCM
Concurrent drugs (drug-drug interactions)	SCM/National Electronic Health Records (NEHR)
Significant co-morbidities/risk factors for resistance	SCM/National Electronic Health Records (NEHR)
Hospital's antibiogram	Laboratory Information System (LIS)
Antibiotic treatment guidelines	SCM/Clinical Decision Support System (CDSS)

Figure 6. Disparate information within various systems

Technical challenges

To tackle data quality issues, numerous datasets need to be categorised and processed. These challenges can be addressed through data engineering and data completeness strategies.

Data engineering transforms medical information collected at various points of the patient's journey into useful components for analysis for the specific illness. For example, chronic illnesses are considered through the patient's lifetime whether the diagnosis was made 1 month ago or 10 years ago. In contrast, lab test results for microbiology will only be considered valid if symptoms are presented within 3 months.

Data-completeness strategies address missing information, which is needed by the model. These could be due to absence of an ordered test by the physician or lack of records. These strategies range from averaging out the missing information or computing a proxy based on other patient medical factors.

Addressing these technical challenges is integral to enabling AI models to be embedded within the hospital system. Once embedded and accessible to the ASP team and doctors, the potential to significantly improve antibiotic prescribing patterns can be realised.

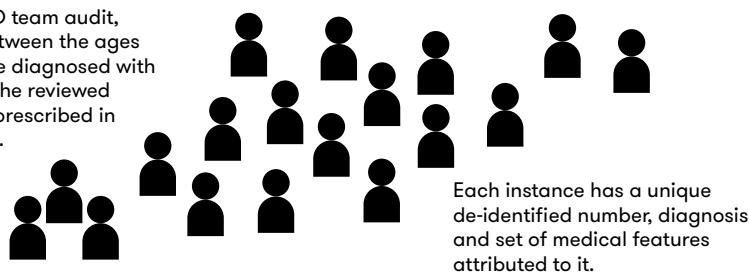
Improving antibiotic stewardship unit oversight

Despite the challenges with data, DXC and SGH have made significant progress with the creation of an AI model for antibiotic analysis.

The first phase of the process has been to build a model from ASP's dataset based on seven broad-spectrum intravenous antibiotics commonly used in the treatment of pneumonia.

Figure 7. Dataset: Clinical samples collected between January 2016 and December 2018

At the time of ASD team audit, **7,421** patients between the ages of **21** and **99** were diagnosed with pneumonia and the reviewed antibiotics were prescribed in **10,459** instances.



Each antibiotic prescription was subjected to the clinical reasoning framework detailed on the page where the five categories of questions were used to determine the presence of an infection, understand the patient and determine previous illnesses and treatments with antibiotics. The data was summarised based on the patient displaying any signs of symptoms within the relevant period. Additional information such as the number of days between patient admissions and administration of the pneumonia regimen was computed and included in the model.

Manual vs. AI

Using AI algorithms, the teams have been able to achieve some important efficiencies, moving the hospital closer to the objective of real-time information on the patient's condition and the most suitable type and method of treatment. Existing standard operational procedures of the ASP involve the manual review of all audited antibiotics for pneumonia, out of which 4 per cent required intervention from the ASP team in the form of alternative antibiotic treatments.

Within this process, there are two key decisions that the ASP team needs to make.

The first is the review of the cases to decide if the treatment is appropriate (operational AI model). The second decision happens when a pneumonia treatment is deemed inappropriate and requiring intervention, with the decision of a suitable regimen of antibiotics to be prescribed (recommender AI model).

The operational AI model identifies treatment profiles that require additional review and can narrow down the number of cases requiring immediate attention by the ASP team. In a test using 2,012 pneumonia treatments unseen by the model, 624 cases were flagged as requiring additional review. Seventy-two (87 per cent) of all intervention cases were captured within the 624 cases flagged for review. Essentially, the model reduces the number of cases for review by a factor of three (624 out of 2,012) and increases the possibility of identifying cases requiring intervention within the prioritised cases by three times — 11.5 per cent — as compared to the full review process, which identifies 4 per cent of all cases for intervention.

The recommender AI model suggests the appropriate antibiotics for each case and can be used to provide guidance to clinicians at the point of prescription. Due to the large number of possible antibiotic regimens, the recommender model was only able to recommend the correct regimen 50 per cent of the time.

Speed of data analysis is also important to the success rate. While manual processes typically take 20 minutes to review each case, AI can assess the same data in less than a second, allowing the ASP teams to assess more data in real time for more antibiotics. This could mean reviewing more cases and expanding the ASP coverage to a larger set of antibiotics.

The next phase

The next goal is to gather more and deeper insights across all antibiotics to treat pneumonia and further enhance operational efficiency. By the third phase, the objective is to have those insights to help treat all infectious diseases with high levels of operational efficiency, providing a platform to support the overarching goal of the following:

To move this objective forward and achieve a comprehensive and powerful AI tool to address antibiotic overuse and misuse, DXC and SGH will need to access a broader data set, including community data.

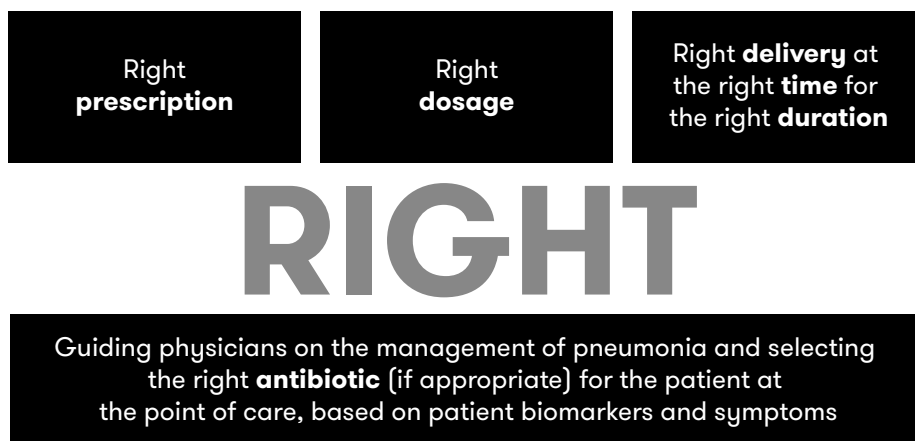


Figure 8. Key goals for the next phase

The goal is to have the automated system integrated with the hospitals' current clinical systems, so prescribing decisions can be supported at the point of care. Consequently, the hospital will be able to prevent inappropriate use of antibiotics and develop more personalised treatment regimens that consider each patient's biomarkers, symptoms, comorbidities and medical history.

Proposed project plan

Value proposition — AI aid for antimicrobial prescription

	Input	Output	Benefit
Phase 1	Limited data set	Appropriate/inappropriate ABX @ ~ 85% accuracy	Test-bed solution
Phase 2	Full data sets (all bio-marker, x-ray, vitals, microbiology) for pneumonia	Appropriate/inappropriate ABX @ higher level of accuracy that triages all normal cases and exception cases routed to ASP human panel	Reduced turnaround time, averted patient time on inappropriate ABX, validated existing guidelines and data-driven approach to new guideline development — ASU
	Pneumonia dose, duration, route; demonstrate framework applicability to UTI	Appropriate ABX for given case	
	Full data set and full interoperability with clinical system	Right ABX at point of care across other IP settings	Wider patient net, ASU role in disease management and cohort identification, validated existing guidelines and data-driven approach to new guideline development
Phase 3		Platform for PoC developed to increase scope of coverage of antibiotics	

Figure 9. An overview of the value proposition for the three phases

Addressing a global health epidemic

The problem of antimicrobial resistance undoubtedly needs a global response. In the Asia-Pacific and beyond, Singapore is recognised as a centre of excellence. Leadership taken in Singapore to tackle the problem through progressive antimicrobial stewardship programs, supported by AI and automation, will pave the way for other markets to adopt similar strategies and improve global antimicrobial stewardship. Furthermore, holistic adoption of such an approach can greatly enhance situational awareness of prescription practices and use of antibiotics.

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